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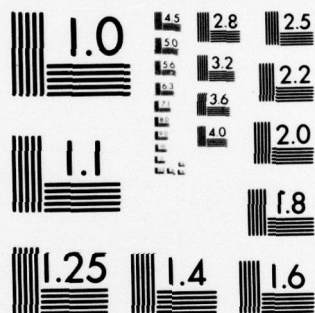
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1. INTRODUCTION

Let $\{X_i, i \geq 1\}$ be a sequence of independent and identically distributed random variables (i.i.d.r.v.) with a distribution function (df) F , defined on $(-\infty, \infty)$. It is assumed that F has a finite (unknown) lower end-point θ , that is

$$(1.1) \quad -\infty < \theta = \sup\{x: F(x) = 0\} < \infty$$

and that $F(x)$ is continuous and monotonic in $x \in (\theta, \theta + \delta)$, for some $\delta > 0$. A natural estimator of θ is the sample minimum i.e.,

$$(1.2) \quad \hat{\theta}_n = \min\{X_1, \dots, X_n\} = X_{n,1} \quad (n \geq 1),$$

where $X_{n,1} \leq \dots \leq X_{n,n}$ stand for the ordered variables corresponding to X_1, \dots, X_n ; $n \geq 1$. $\hat{\theta}_n$ is a (strongly) consistent estimator of θ , but it is not an unbiased one; the nature of its bias depends on the order of terminal contact of F (at θ). It may therefore be appealing to use the jackknife estimator corresponding to $\hat{\theta}_n$.

Under quite general regularity conditions (viz., [1,2,4]), jackknifing meets three objectives: (a) *Bias reduction*. If θ_n^* be the jackknife estimator then $nE(\theta_n^* - \theta) \rightarrow 0$ as $n \rightarrow \infty$. (b) *Asymptotic normality*. If $n^{1/2}(\hat{\theta}_n - \theta)$ is asymptotically normal, then the same limit law holds for $n^{1/2}(\theta_n^* - \theta)$. (c) The Tukey estimator v_n^2 [defined by (2.5)] is a (strongly) consistent estimator of the variance of $n^{1/2}(\theta_n^* - \theta)$.

Since the asymptotic distributions of sample extrema are non-normal and, depending on the order of terminal contact, the bias of $\hat{\theta}_n$ is $O(n^{-a})$

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for some $0 < a \leq 1$, the effectiveness of jackknifing in regard to (a) and (b) remains to be examined carefully. Further, in this case, v_n^2 does not converge (stochastically). Along with the preliminary notions, expressions for θ_n^* and v_n^2 are considered in Section 2. The main results are studied in Section 3. Section 4 deals with a modification of jackknifing appropriate for the case of the bias of $O(n^{-a})$ for some $a < 1$. Some general remarks are made in the concluding section.

2. PRELIMINARY NOTIONS

We assume that for some non-negative integer m , $F(x)$ has continuous j th derivative $F^{(j)}(x) [= f^{(j-1)}(x)]$ for all $x \in (\theta, \theta + \delta)$, $\delta > 0$, $1 \leq j \leq m+1$. We denote the (right hand) derivatives at θ by $F_+^{(j)}(\theta) = f_+^{(j-1)}(\theta)$, $1 \leq j \leq m+1$ and $F_+^{(0)}(\theta) = 0$, $f_+^{(0)}(\theta) = f_+(\theta)$. Then, a *terminal contact of order m* is defined by

$$(2.1) \quad F_+^{(j)}(\theta) = 0, \quad 0 \leq j \leq m \quad \text{and} \quad 0 < f_+^{(m)}(\theta) < \infty.$$

Also, for the study of the bias, we assume that

$$(2.2) \quad v_\alpha = \int_0^\infty |x|^\alpha dF(x) < \infty \quad \text{for some } \alpha > 0.$$

To define θ_n^* , we let for each i : $1 \leq i \leq n$,

$$(2.3) \quad \hat{\theta}_{n-1}^i = \min\{X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_n\}, \quad \hat{\theta}_{n,i} = n\hat{\theta}_n - (n-1)\hat{\theta}_{n-1}^i.$$

Then, $\hat{\theta}_{n-1}^i$ is equal to $X_{n,1}$ or $X_{n,2}$ according as X_i is \neq or $= X_{n,1}$, $1 \leq i \leq n$. Also,

$$(2.4) \quad \begin{aligned} \theta_n^* &= n^{-1} \sum_{i=1}^n \hat{\theta}_{n,i} \\ &= X_{n,1} - n^{-1}(n-1)(X_{n,2} - X_{n,1}), \quad n \geq 2. \end{aligned}$$

The Tukey estimator v_n^2 , defined by

$$(2.5) \quad v_n^2 = \frac{1}{(n-1)} \sum_{i=1}^n (\hat{\theta}_{n,i} - \theta_n^*)^2 = (n-1) \sum_{i=1}^n (\hat{\theta}_{n-1}^i - \theta_n^*)^2,$$

reduces in our case to

$$(2.6) \quad v_n^2 = (X_{n,2} - X_{n,1})^2 (n-1)(n^2+n-1)/n \quad (\sim \{n(X_{n,2} - X_{n,1})\}^2).$$

For a terminal contact of order $m(\geq 0)$, we define

$$(2.7) \quad b_{n,m} = \{nf_+^{(m)}(\theta)/(m+1)!\}^{1/(m+1)}, \quad a_m = 1/(m+1).$$

Then, the limiting distribution of $b_{n,m}(\hat{\theta}_n - \theta)$ is known to be

$$(2.8) \quad \Lambda_m(x) = \begin{cases} 0, & x \leq 0, \\ 1 - \exp(-x^{m+1}), & x > 0. \end{cases}$$

Also, by Theorem 3.1 of Sen (1961), as $n \rightarrow \infty$,

$$(2.9) \quad b_{n,m} E(X_{n,r} - \theta) = \sqrt{r+a_m}/\sqrt{r} + o(1), \quad \text{for every (fixed) } r(\geq 1).$$

3. BASIC PROPERTIES OF JACKKNIFING

It follows from (2.4) that

$$(3.1) \quad \begin{aligned} n(\theta_n^* - \theta) &= n(X_{n,1} - \theta) - (n-1)(X_{n,2} - X_{n,1}) \\ &= (2n-1)(X_{n,1} - \theta) - (n-1)(X_{n,2} - \theta). \end{aligned}$$

Hence, from (2.9) and (3.1), we obtain that for a terminal contact of order m ,

$$(3.2) \quad \begin{aligned} b_{n,m} E(\theta_n^* - \theta) &= (1-a_m) \sqrt{1+a_m} + o(1) \\ &= (1-a_m) \{b_{n,m} E(\hat{\theta}_n - \theta)\} + o(1) . \end{aligned}$$

For $m=0$ i.e., $a_m=1$, the right hand side (rhs) of (3.2) converges to 0, as $n \rightarrow \infty$, while for $m \geq 1$ (i.e., $a_m \leq \frac{1}{2}$), jackknifing leads to effectively $100(1-a_m)\%$ reduction in bias. Thus, the basic role of jackknifing is partially impaired for a terminal contact of order $m(\geq 1)$.

Theorem 1. For a terminal contact of order $m(\geq 0)$,

$$(3.3) \quad \begin{aligned} \Lambda_n^*(x) &= \lim_{n \rightarrow \infty} P\{b_{n,m}(\theta_n^* - \theta) \leq x\} \\ &= \begin{cases} \int_0^\infty \exp\left\{-(2y^{a_m} - x)^{m+1}\right\} dy, & -\infty < x \leq 0, \\ 1 - \exp(-x^{m+1}) + \int_{x^{m+1}}^\infty \exp\left\{-(2y^{a_m} - x)^{m+1}\right\} dy, & x > 0, \end{cases} \end{aligned}$$

where a_m and $b_{n,m}$ are defined by (2.7).

Proof. Let $Z_n = b_{n,m}(\theta_n^* - \theta)$ and let

$$(3.4) \quad Y_{n(1)} = nF(X_{n,1}) \quad \text{and} \quad Y_{n(2)} = n[F(X_{n,2}) - F(X_{n,1})] .$$

Then, by (2.1), (2.2), (2.7), (3.1) and (3.4) and proceeding as in the proof of Theorem 3.1 of Sen (1961), we obtain that

$$(3.5) \quad E[Z_n - 2Y_{n(1)}^{a_m} + (Y_{n(1)} + Y_{n(2)})^{a_m}]^2 \rightarrow 0 \quad \text{as } n \rightarrow \infty .$$

and hence, by the Chebychev inequality, we have

$$(3.6) \quad \Lambda_m^*(x) = \lim_{n \rightarrow \infty} P \left\{ 2Y_{n(1)}^a - (Y_{n(1)} + Y_{n(2)})^a \leq x \right\}, \quad -\infty < x < \infty.$$

We may recall that $Y_{n(1)}$ and $Y_{n(2)}$ are asymptotically independently distributed according to a common simple exponential law and they are non-negative rv's. For $x \leq 0$, $\left[2Y_{n(1)}^a - (Y_{n(1)} + Y_{n(2)})^a \leq x \right] \Leftrightarrow \left[Y_{n(2)} \geq \left(2Y_{n(1)}^a - x \right)^{1/a} - Y_{n(1)} \right]$ and the first equation in (3.3) follows directly by finding the conditional probability given $Y_{n(1)}$ and then integrating it out over $Y_{n(1)}$. For $x > 0$, if $Y_{n(1)} \leq x^{1/a}$, then $2Y_{n(1)}^a - (Y_{n(1)} + Y_{n(2)})^a \leq Y_{n(1)}^a \leq x$, while for $Y_{n(1)} > x^{1/a}$, as before we need $Y_{n(2)} \geq (2Y_{n(1)}^a - x)^{1/a} - Y_{n(1)}$, and hence, the last equation in (3.3) follows on parallel lines. Q.E.D.

For $m=0$ (i.e., $a_m=1$), Λ_0 in (2.8) is the simple exponential while Λ_0^* in (3.3) is the double exponential df. For $m \geq 0$, Λ_m and Λ_m^* are not the same df.

Theorem 2. For a terminal contact of order $m(\geq 0)$,

$$(3.7) \quad \lim_{n \rightarrow \infty} \left\{ E \left[b_{n,m}^2 (\hat{\theta}_n - \theta)^2 \right] \right\} = \left\{ 1 - \frac{2a_m(1-a_m)}{1+a_m} \right\} \left[\lim_{n \rightarrow \infty} \left\{ E \left[b_{n,m}^2 (\hat{\theta}_n - \theta)^2 \right] \right\} \right] \\ = (2a_m \sqrt{2a_m} \{1 - 2a_m(1-a_m)/(1+a_m)\}) .$$

Proof. Since $\hat{\theta}_n = X_{n,1}$, by an appeal to Theorem 3.1 of Sen (1961), we get that

$$(3.8) \quad b_{n,m}^2 E(\hat{\theta}_n - \theta)^2 \rightarrow \sqrt{1+2a_m} = 2a_m \sqrt{2a_m} > 0 .$$

Hence, to prove (3.7), by (3.5), it suffices to show that as $n \rightarrow \infty$,

$$(3.9) \quad E \left[2Y_{n(1)}^{a_m} - (Y_{n(1)} + Y_{n(2)})^{a_m} \right]^2 \rightarrow 2a_m \sqrt{2a_m} \{1 - 2a_m(1-a_m)/(1+a_m)\}.$$

Towards this, we may note that $E \left[Y_{n(1)}^{2a_m} \right] \sim \sqrt{1+2a_m} = 2a_m \sqrt{2a_m}$, $E(Y_{n(1)} + Y_{n(2)})^{2a_m} \rightarrow \sqrt{2+2a_m} = 2a_m(1+2a_m)\sqrt{2a_m}$ while $E \left\{ Y_{n(1)}^{a_m} (Y_{n(1)} + Y_{n(2)})^{a_m} \right\} = E \left\{ \left[E(Y_{n(1)}^{a_m} | Y_{n(1)} + Y_{n(2)}) \right] (Y_{n(1)} + Y_{n(2)})^{a_m} \right\} \sim E \left\{ (Y_{n(1)} + Y_{n(2)})^{2a_m+1} / (a_m+1) \right\} \rightarrow \left(\sqrt{2+2a_m} \right) / (1+a_m) = \left(2a_m(1+2a_m)\sqrt{2a_m} \right) / (a_m+1)$, and hence, (3.9) follows by some standard steps. Q.E.D.

For $m = 0$ (i.e., $a_m = 1$), the second factor on the rhs of (3.7) is equal to 1, so that both $\hat{\theta}_n$ and θ_n^* have the same asymptotic variance, though their df's are not the same. For $m \geq 1$ (i.e., $a_m \leq 1/2$), $2a_m(1-a_m)/(1+a_m) > 0$ and is bounded from above by $1/3$. Thus, from (3.2) and (3.7) we have that jackknifing reduces both the asymptotic bias and the asymptotic mean square to a fractional extent. This characteristic is different from the regular case where there is a complete reduction of asymptotic bias but no reduction of the asymptotic mean square.

From (2.6), (2.7) and (3.4), it follows that for a terminal contact of order $m(\geq 0)$,

$$(3.10) \quad \left| n^{-1}b_{n,m}v_n - \left\{ (Y_{n(1)} + Y_{n(2)})^{a_m} - Y_{n(1)}^{a_m} \right\} \right| \xrightarrow{p} 0, \text{ as } n \rightarrow \infty.$$

Since $(Y_{n(1)} + Y_{n(2)})^{a_m} - Y_{n(1)}^{a_m} \rightarrow \left\{ (Y_1 + Y_2)^{a_m} - Y_1^{a_m} \right\}$, where Y_1 and Y_2 are i.i.d.r.v. having the simple exponential df on $[0, \infty)$, $n^{-1}b_{n,m}$ either converges to a positive constant (when $m = 0$) or goes to 0 (when $m \geq 1$), it follows that either (for $m = 0$) v_n has a non-degenerate asymptotic df

or (for $m \geq 1$) it goes to $+\infty$, in probability as $n \rightarrow \infty$. This characteristic is also different from the regular case where $v_n \xrightarrow{P} a$ constant, as $n \rightarrow \infty$. Nevertheless, for the studentized form, we have for a terminal contact of order $m(\geq 0)$,

$$(3.11) \quad T_n = n(\theta_n^* - \theta)/v_n = b_{n,m}(X_{n,1} - \theta)/b_{n,m}(X_{n,2} - X_{n,1}) - (n-1)/n \\ + o_p(1) \xrightarrow{P} Y_1^a / \left\{ (Y_1 + Y_2)^a - Y_1^a \right\} - 1,$$

so that noting that $Y^* = Y_2/Y_1$ has the Fisher's variance-ratio distribution with degrees of freedom (2,2), we have from (3.11) that

$$(3.12) \quad [1 + (1 + T_n)^{-1}]^{m+1} - 1 \xrightarrow{P} Y^* = Y_2/Y_1.$$

For $m = 0$, we have a simplified form

$$(3.13) \quad T_n + 1 \xrightarrow{P} Y_1/Y_2 \stackrel{P}{=} Y^*.$$

Both (3.12) and (3.13) have important statistical applications.

4. A MODIFICATION OF θ_n^*

We have observed in (3.2) that for $m \geq 1$, $b_{n,m}E(\theta_n^* - \theta)$ does not converge to 0 as $n \rightarrow \infty$. Let C_n be the sigma-field generated by $X_{n,1}, \dots, X_{n,n}$ and by X_{n+j} , $j \geq 1$ (so that C_n is non-increasing in n). Then, in the regular case, [cf. (2.11) of Sen (1977)], we have

$$(4.1) \quad \theta_n^* - \hat{\theta}_n = (n-1)E\{(\hat{\theta}_n - \hat{\theta}_{n-1})|C_n\}.$$

In our case, for $m \geq 1$, $nb_{n,m}E(\hat{\theta}_n - \hat{\theta}_{n-1}) = -a_m \sqrt{1+a_m} + o(1)$, where as $b_{n,m}E(\hat{\theta}_n - \theta) = \sqrt{1+a_m} + o(1)$, and thereby, we get the resulting bias in (3.2). To eliminate the, we may consider the modified estimator

$$(4.2) \quad \begin{aligned} \theta_{n,m}^{**} &= \hat{\theta}_n + \frac{1}{a_m} E\{(\hat{\theta}_n - \hat{\theta}_{n-1}) | C_n\} \\ &= X_{n,1} - (m+1)n^{-1}(n-1)(X_{n,2} - X_{n,1}) . \end{aligned}$$

In that case, we have

$$(4.3) \quad b_{n,m}E(\theta_{n,m}^{**} - \theta) \rightarrow 0 \quad \text{as } n \rightarrow \infty .$$

Also, following the same line as in the proof of Theorem 1, we obtain that

$$(4.4) \quad \begin{aligned} \Lambda_m^{**}(x) &= \lim_{n \rightarrow \infty} P\{b_{n,m}(\theta_{n,m}^{**} - \theta) \leq x\} \\ &= \begin{cases} \int_0^{\infty} \exp\left\{-\left[\frac{(m+2)y^m - x}{(m+1)}\right]^{m+1}\right\} dy, & -\infty < x \leq 0, \\ 1 - \exp\{-x^{m+1}\} + \int_x^{\infty} \exp\left\{-\left[\frac{(m+2)y^m - x}{(m+1)}\right]^{m+1}\right\} dy, & 0 < x < \infty. \end{cases} \end{aligned}$$

Also, following the line of proof of Theorem 2, we have

$$(4.5) \quad \begin{aligned} \lim_{n \rightarrow \infty} E\{b_{n,m}^2(\theta_{n,m}^{**} - \theta)^2\} &= (2a_m \sqrt{2a_m}) \left\{1 - \frac{2a_m}{1+a_m} (m+1)[(m+1)a_m - 1]\right\} = 2a_m \sqrt{2a_m} \\ &= \lim_{n \rightarrow \infty} E\{b_{n,m}^2(\hat{\theta}_n - \theta)^2\} \geq \lim_{n \rightarrow \infty} E\{b_{n,m}^2(\theta_n^* - \theta)^2\} . \end{aligned}$$

Thus, whereas $\theta_{n,m}^{**}$ eliminates bias to the desired extent, it fails to reduce the mean square. In this sense, it is similar to the case of θ_n^* in the regular case. [Though Λ_m^{**} and Λ_m are not the same.]

Finally, for the studentized case, in (3.11)-(3.13), the only changes we need to make is to replace T_n by $T_n + m$; the rest remains the same.

5. SOME REMARKS

We have so far considered the case of the lower end-point. The case of the upper end-point (if finite) follows on parallel lines. Secondly, in practical applications, when the form of F is not specified but the order of terminal contact is assumed to be known [viz., $m = 0$ when F is U-shaped or inverted J-shaped, etc.], the studentized form in (3.11)-(3.13) may most conveniently be used to provide a jackknife test for a null hypothesis $H_0: \theta = \theta_0$ (specified) or a confidence interval for the unknown θ . For a symmetric df with both end-points finite, jackknifing of the extreme mid-range (for estimating or testing for the location of the df) can be made — the jackknife estimator corresponding to the smallest and the largest order statistic are also asymptotically independent.

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